# PLS Mean Centering

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# Partial Least Squares (PLS) with Mean Centering

Partial least squares usually uses mean-centered data to compute  $C_{\rm pls}$  using the weights from the SIMPLS algorithm with r latent variables.

```
X = A_train;
Y = C_train;
X0 = X - mean(X,1);
Y0 = Y - mean(Y,1);
[X_loadings, Y_loadings, X_scores, Y_scores, Weights] = pnnl_simpls(X0, Y0, r);
B_pls = Weights * Y_loadings';
C_pls = (A_unknown - mean(A_train,1)) * B_pls + mean(C_train,1);
```

# Partial Least Squares (PLS) without Mean Centering

The following is PLS without mean centering.

```
X = A_train;
Y = C_train;
[X_loadings,Y_loadings,X_scores,Y_scores,Weights] = pnnl_simpls(X,Y,r);
B_pls = Weights * Y_loadings';
C_pls = A_unknown * B_pls;
```

# **Combined algorithm**

To make it easier to run with the rest of the tools in the PNNL toolbox, we combined mean-centered and non-mean-centered into one function with meanCentered as an optional argument. When meanCentered is not used as an input, then the default is to compute without mean-centered data.

```
function [C_pls, B_pls] = pnnl_pls(A_train, C_train, A_unknown, r, meanCentered)
    if nargin < 5
        meanCentered = false;
end

X = A_train;
Y = C_train;
if meanCentered
        X0 = X - mean(X,1);
        Y0 = Y - mean(Y,1);
else
        X0 = X;
        Y0 = Y;
end

[X_loadings,Y_loadings,X_scores,Y_scores,Weights] = pnnl_simpls(X0,Y0,r); %#ok<ASGLU>
```

```
B_pls = Weights * Y_loadings';

if meanCentered
    C_pls = (A_unknown - mean(A_train,1)) * B_pls + mean(C_train,1);
else
    C_pls = A_unknown * B_pls;
end
end
```

## **Napalm Data**

Load the included napalm data to run the PLS algorithms.

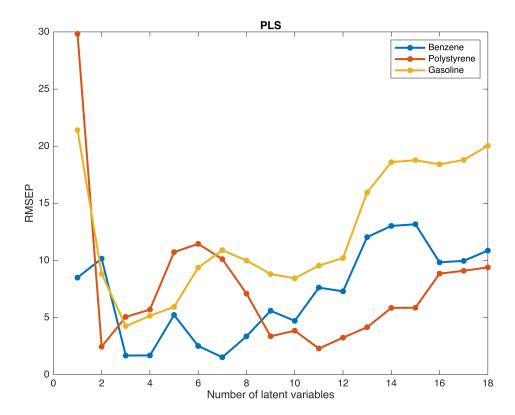
```
clearvars
load pnnl_napalm_data
whos
```

Name	Size	Bytes	Class	Attributes
A_train	20x1713	274080	double	
A_unknown	12x1713	164448	double	
C_train	20x3	480	double	
<pre>C_validation</pre>	12x3	288	double	
ConcentrationUnits	1x4	8	char	
ConstituentNames	1x3	364	cell	
WavenumberLabel	1x20	40	char	
Wavenumbers	1x1713	13704	double	
ans	1x1	8	double	

# Optimal number of latent variables for PLS with mean centering

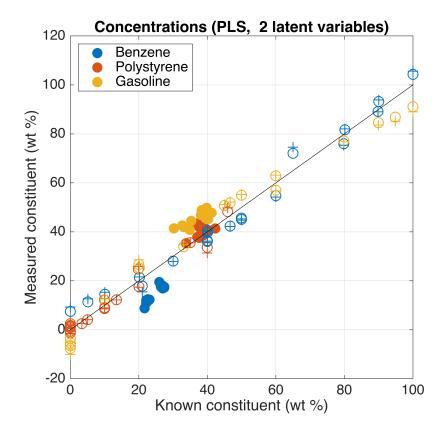
Compute PLS with mean centering for 1 through 18 latent variables and plot RMSEP for them.

```
nLatentVariables = 1:18;
meanCentered = true;
[C_pls, RMSEP_pls] = pnnl_napalm_pls(nLatentVariables, meanCentered);
plot(nLatentVariables, RMSEP_pls,'.-','LineWidth',2,'MarkerSize',20)
xlabel('Number of latent variables')
ylabel('RMSEP')
title('PLS')
legend(ConstituentNames{:})
```



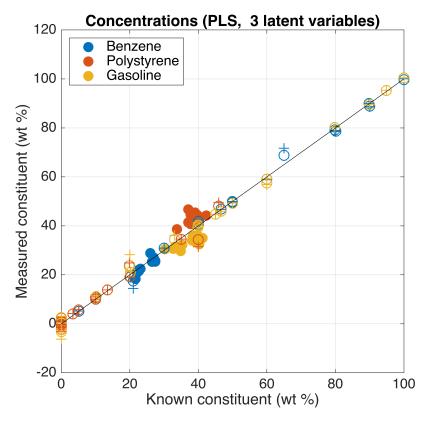
For mean-centered data, it looks like the knee in the curve for polystyrene is 2 latent variables, and 3 for benzene and gasoline. Plot them to see what they look like.

```
nLatentVariables = [2,3];
pnnl_napalm_pls(nLatentVariables, meanCentered);
```



Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS,	2 latent variables	Benzene	Polystyrene	Gasoline
	RMSEC	4.1479	2.296	5.5014
	RMSECV	5.0229	2.924	6.5258
	RMSEP	10.164	2.4566	8.8266
	KI'15EP 	10.104	2.4300 	8.8200



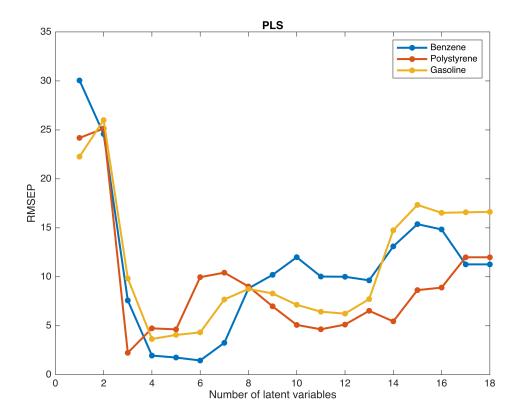
Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS,	3 latent variables	Benzene	Polystyrene	Gasoline
	RMSEC	1.3968	1.8465	1.5789
	RMSECV	2.2406	2.6176	2.7657
	RMSEP	1.6799	5.0647	4.2528

# Optimal number of latent variables for PLS without mean omentering

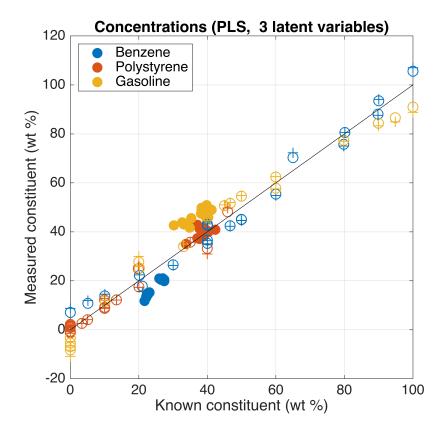
Compute PLS witout mean centering for 1 through 18 latent variables and plot RMSEP for them.

```
nLatentVariables = 1:18;
meanCentered = false;
[C_pls, RMSEP_pls] = pnnl_napalm_pls(nLatentVariables, meanCentered);
plot(nLatentVariables, RMSEP_pls,'.-','LineWidth',2,'MarkerSize',20)
xlabel('Number of latent variables')
ylabel('RMSEP')
title('PLS')
legend(ConstituentNames{:})
```



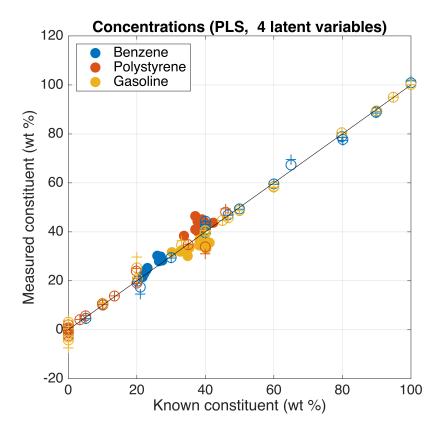
Without mean centering, it looks like the knee in the curve for polystyrene is 3 latent variables, 4 for gasoline, and 6 for benzene. Plot them to see what they look like.

```
nLatentVariables = [3 4 6];
pnnl_napalm_pls(nLatentVariables, meanCentered);
```



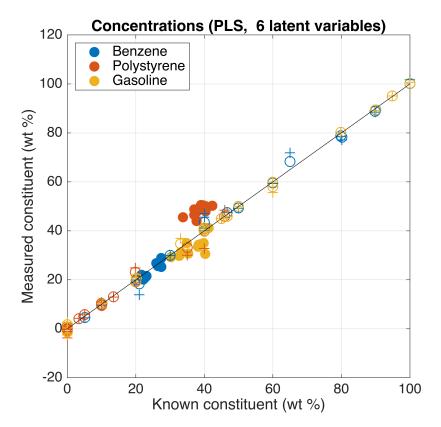
Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS,	3 latent variables	Benzene	Polystyrene	Gasoline
	RMSEC	4.1853	2.3592	5.6109
	RMSECV	5.0232	2.9897	6.6976
	RMSEP	7.5851	2.2404	9.8369



Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS,	Ri	ables RMSEC MSECV RMSEP	Benzene 1.6636 2.3112 1.9599	Polystyrene 1.921 2.6884 4.7379	1.9509 3.2095



Legend: Dot is predicted. Circle is train. Cross is cross-validation.

PLS,	6 latent variables	Benzene	Polystyrene	Gasoline
	RMSEC	1.4134	0.99256	0.91242
	RMSECV	3.1544	2.5526	1.7502
	RMSEP	1.4567	9.9656	4.3211

### %#ok<\*ASGLU>

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